Semi-generic analysis of Kraken results from Sunbeam

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Required input:

- Kraken2 output from Sunbeam (demo_kraken_all_samples.tsv)
- Sample metadata (could also construct it from the sample names in the Kraken dataset)

Analyses we will perform:

- Alpha diversity
- Barplots of microbial composition
- Beta diversity AKA PCoA
- DESeq2 to identify differential taxa

Search the document for "CHANGEME" to see which lines of code will require project-specific tweaks.

Load libraries

```
library(tidyverse)
library(phyloseq)
library(DESeq2)

# speedyseq is just to speed up tax_glom
# remotes::install_github("mikemc/speedyseq")
library(speedyseq)

theme_set(theme_bw(base_size=15))
```

Import data

```
dim(kraken.data.df)
## [1] 6565
               16
head(kraken.data.df)
           CR40.2002_044w CR40.2004_044w CR40.2005_044w CR40.2006_044w
##
## 2
                      7953
                                      18066
                                                      22056
                                                                       10786
## 1239
                       4224
                                      15247
                                                       5666
                                                                        4349
## 186801
                       154
                                        564
                                                        143
                                                                         157
## 186802
                       6016
                                      21858
                                                       5837
                                                                        6674
                                                                        2407
## 186803
                       2582
                                      10317
                                                       2453
## 1506553
                        129
                                        733
                                                        130
                                                                         188
            CR40.2007_044w CR40.2008_044w CR40.2010_044w AL.0004_044w AL.0005_044w
##
## 2
                     23040
                                      15772
                                                      27846
                                                                    11038
                                                                                    6493
                                                                                    6130
## 1239
                     23822
                                      11088
                                                       8162
                                                                     3296
## 186801
                       888
                                        373
                                                        216
                                                                      133
                                                                                     207
## 186802
                                      16897
                                                      11008
                                                                     4258
                                                                                  10365
                     35103
                     14085
                                                                     1695
## 186803
                                       6776
                                                       4081
                                                                                    4082
## 1506553
                        796
                                        494
                                                        234
                                                                        90
                                                                                     225
##
            AL.0006_044w AL.0007_044w AL.0008_044w AL.0010_044w AL.0012_044w
## 2
                    2588
                                 13389
                                                 7739
                                                              15706
                                                                             8642
## 1239
                    1941
                                 13640
                                                10183
                                                              13891
                                                                            11088
## 186801
                      73
                                    462
                                                  262
                                                                508
                                                                              365
## 186802
                    2796
                                 21574
                                                14860
                                                              21257
                                                                            18023
## 186803
                    1386
                                  7990
                                                5573
                                                              10053
                                                                             8115
## 1506553
                      56
                                    523
                                                                504
                                                                              514
                                                  519
##
            AL.0014 044w AL.0015 044w
## 2
                   14922
                                 23453
## 1239
                   13761
                                  8700
## 186801
                     651
                                    295
## 186802
                                 10149
                   27305
## 186803
                                  3840
                   10761
## 1506553
                                    220
                     601
```

- 6565 taxa, 16 samples
- Make sure these numbers (especially the number of samples) make sense

Import sample metadata

```
# CHANGEME: you can either create sample metadata from the sample names
# or import the metadata from a different file
meta.df <- data.frame(
    sample.ID=colnames(kraken.data.df))

# CHANGEME: create new columns based on sample name
meta.df <- meta.df %>%
    tidyr::separate(sample.ID, c("diet.mouseID", "age.weeks"), sep="_", remove=F) %>%
    tidyr::separate(diet.mouseID, c("diet", "mouseID"), sep="\\.") %>%
    column_to_rownames("sample.ID")

# CHANGEME: convert certain columns to factors in order to specify their order in plots
meta.df <- meta.df %>%
    mutate(diet.fctr=factor(diet, levels=c("AL", "CR40")))
```

Convert Consensus Lineage column into a taxonomy df

```
kraken.tax.df <- data.frame(str_split_fixed(kraken.df$Consensus.Lineage, "__|; ", n=14)[, seq(2,14,2)])
colnames(kraken.tax.df) <- c("kingdom", "phylum", "class", "order", "family", "genus", "species")
rownames(kraken.tax.df) <- kraken.df$OTU.ID</pre>
```

Create phyloseq object

```
physeq <- phyloseq(
  kraken.data.df %>% as.matrix %>% otu_table(taxa_are_rows=T),
  meta.df %>% sample_data,
  kraken.tax.df %>% as.matrix %>% tax_table)
```

Subset to bacteria

```
physeq.bac <- physeq %>% subset_taxa(kingdom=="Bacteria")
```

• 6242 bacterial taxa

Percent of reads assigned to each taxonomy level

```
physeq.bac.tax.df.w.tax.sum <- data.frame(taxa.sum=taxa_sums(physeq.bac)) %>%
    merge(data.frame(tax_table(physeq.bac)), by="row.names")

1 - sum(physeq.bac.tax.df.w.tax.sum$taxa.sum[physeq.bac.tax.df.w.tax.sum$family == ""]) / sum(physeq.bac.tax.df.w.tax.sum$family == ""]) / sum(physeq.bac.tax.df.w.tax.sum$genus == ""]) / sum(physeq.bac.tax.df.w.tax.sum$genus == ""]) / sum(physeq.bac.tax.df.w.tax.sum$physeq.bac.tax.df.w.tax.sum$species == ""]) / sum(physeq.bac.tax.df.w.tax.sum$species == ""]) / sum(physeq.bac.tax.df.w.tax.sum$specie
```

- These numbers describe what percent of reads could be mapped to a particular taxonomy level
- For example, 92.7% of reads could be mapped at least to the genus level

Aggregate to different taxonomy levels

```
physeq.bac.phylum <- physeq.bac %>% tax_glom(taxrank="phylum")
physeq.bac.class <- physeq.bac %>% tax_glom(taxrank="class")
physeq.bac.family <- physeq.bac %>% tax_glom(taxrank="family")
physeq.bac.genus <- physeq.bac %>% tax_glom(taxrank="genus")
physeq.bac.species <- physeq.bac %>% tax_glom(taxrank="species")
ntaxa(physeq.bac.phylum)
## [1] 37
ntaxa(physeq.bac.class)
```

```
## [1] 73
ntaxa(physeq.bac.family)
## [1] 384
ntaxa(physeq.bac.genus)
## [1] 1463
ntaxa(physeq.bac.species)
```

[1] 5487

• This tells you how many unique bacterial phyla, families, genera, etc. were detected

Add genus_and_species annotation to physeq.bac.species for plotting purposes

```
tax_table(physeq.bac.species) <- data.frame(tax_table(physeq.bac.species)) %>%
  mutate(genus_and_species=str_c(genus, species, sep=" ")) %>%
 as.matrix %>% tax table
```

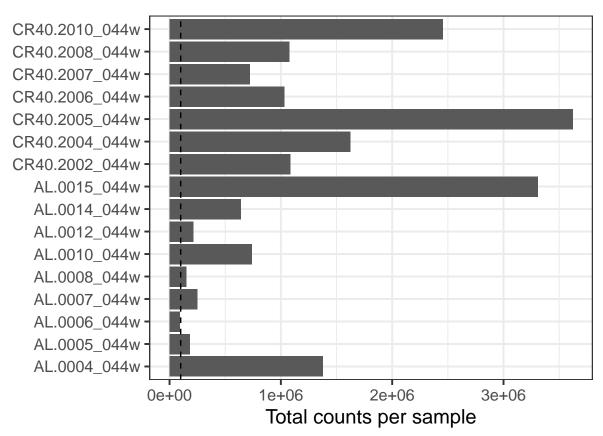
Optional: Filtration

Samples

It may be helpful to discard samples with very few reads and taxons that were observed very infrequently. The defaults provided here are to include samples with at least 10k counts, and taxons with at least 0.01% relative abundance in 20% of samples.

Here, I perform filtering for just the genus phyloseq object, but could do the same thing for the phyloseq objects at other taxonomy levels.

```
# CHANGEME: update filtering parameters as desired
MIN.COUNTS.PER.SAMPLE <- 1e5
data.frame(sample_sums(physeq.bac)) %>%
  setNames("sample.sum") %>%
  rownames to column("sample.ID") %>%
  ggplot(aes(x=sample.sum, y=sample.ID)) +
  geom_bar(stat="identity") +
  geom_vline(xintercept=MIN.COUNTS.PER.SAMPLE, lty=2) +
  labs(x="Total counts per sample", y="")
```



```
samples.to.keep.bool <- sample_sums(physeq.bac) > MIN.COUNTS.PER.SAMPLE

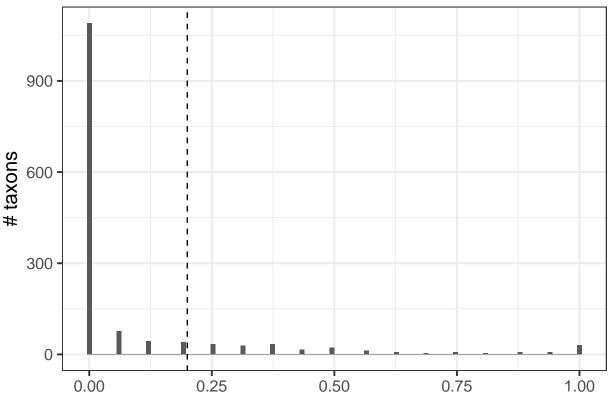
# CHANGEME: apply filtering for all phyloseq objects that you want
physeq.bac.genus.filt <- physeq.bac.genus %>%
    subset_samples(samples.to.keep.bool)
physeq.bac.species.filt <- physeq.bac.species %>%
    subset_samples(samples.to.keep.bool)
```

Taxa

Can filter the phyloseg objects at other taxonomy levels too.

```
# CHANGEME: update filtering parameters as desired
MIN.REL.ABUND <- 1e-4
MIN.PERC.SAMPLES.W.TAXON.AT.REL.ABUND <- 0.2

bac.genus.rel.abund.mat <- physeq.bac.genus %>%
    transform_sample_counts(function(OTU) OTU/sum(OTU)) %>%
    otu_table %>% as.matrix
data.frame(
    frac.taxon.w.sufficient.rel.abund=rowSums(bac.genus.rel.abund.mat > MIN.REL.ABUND) / ncol(bac.genus.r
    ggplot(aes(frac.taxon.w.sufficient.rel.abund)) +
    geom_histogram(bins=100) +
    geom_vline(xintercept=MIN.PERC.SAMPLES.W.TAXON.AT.REL.ABUND, lty=2) +
    labs(y="# taxons", x=sprintf("Fraction of samples in which a taxon had relative abundance > %.3f%%", it
```



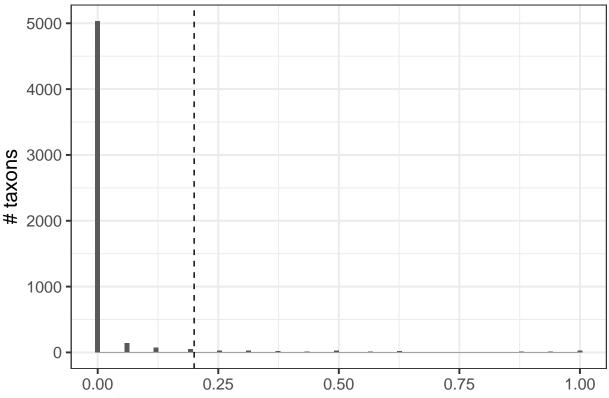
Fraction of samples in which a taxon had relative abundance > 0.01

```
genera.to.keep.bool <- rowSums(bac.genus.rel.abund.mat > MIN.REL.ABUND) / ncol(bac.genus.rel.abund.mat)
physeq.bac.genus.filt <- physeq.bac.genus.filt %>%
   subset_taxa(genera.to.keep.bool)
```

Species

```
# CHANGEME: update filtering parameters as desired
MIN.REL.ABUND <- 1e-4
MIN.PERC.SAMPLES.W.TAXON.AT.REL.ABUND <- 0.2

bac.species.rel.abund.mat <- physeq.bac.species %>%
    transform_sample_counts(function(OTU) OTU/sum(OTU)) %>%
    otu_table %>% as.matrix
data.frame(
    frac.taxon.w.sufficient.rel.abund=rowSums(bac.species.rel.abund.mat > MIN.REL.ABUND) / ncol(bac.speci
ggplot(aes(frac.taxon.w.sufficient.rel.abund)) +
    geom_histogram(bins=100) +
    geom_vline(xintercept=MIN.PERC.SAMPLES.W.TAXON.AT.REL.ABUND, lty=2) +
    labs(y="# taxons", x=sprintf("Fraction of samples in which a taxon had relative abundance > %.3f%", it
```



Fraction of samples in which a taxon had relative abundance > 0.0°

```
species.to.keep.bool <- rowSums(bac.species.rel.abund.mat > MIN.REL.ABUND) / ncol(bac.species.rel.abund
physeq.bac.species.filt <- physeq.bac.species.filt %>%
   subset_taxa(species.to.keep.bool)
```

Summary of filtration

```
physeq.bac.genus
## phyloseq-class experiment-level object
## otu_table()
                OTU Table:
                              [ 1463 taxa and 16 samples ]:
## sample_data() Sample Data:
                                    [ 16 samples by 4 sample variables ]:
## tax_table()
                Taxonomy Table:
                                    [ 1463 taxa by 7 taxonomic ranks ]:
## taxa are rows
physeq.bac.genus.filt
## phyloseq-class experiment-level object
## otu_table()
                OTU Table:
                                    [ 213 taxa and 15 samples ]:
                                    [ 15 samples by 4 sample variables ]:
## sample_data() Sample Data:
## tax table()
                Taxonomy Table:
                                    [ 213 taxa by 7 taxonomic ranks ]:
## taxa are rows
physeq.bac.species
## phyloseq-class experiment-level object
## otu_table()
                OTU Table:
                                    [ 5487 taxa and 16 samples ]:
                                    [ 16 samples by 4 sample variables ]:
## sample_data() Sample Data:
                                   [ 5487 taxa by 8 taxonomic ranks ]:
## tax_table()
                Taxonomy Table:
## taxa are rows
```

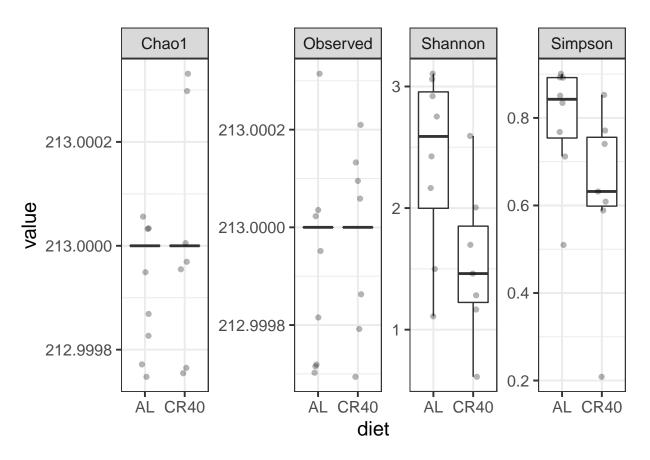
- $16 \text{ samples} \rightarrow 15 \text{ samples}$
- 1463 genera -> 213 genera
- 5487 species -> 197 species

Alpha diversity

```
alpha.div.genus.df <- estimate_richness(
   physeq.bac.genus.filt, measures=c("Observed", "Chao1", "Shannon", "Simpson"))

alpha.div.genus.df %>%
   merge(meta.df, by="row.names") %>%
   pivot_longer(c(Observed, Chao1, Shannon, Simpson), names_to="diversity") %>%

# CHANGEME to show desired comparisons
ggplot(aes(x=diet, y=value)) +
geom_boxplot(outlier.shape=NA) +
geom_jitter(width=0.1, alpha=0.3) +
facet_wrap(~diversity, ncol=4, scales="free_y")
```



After rarefaction

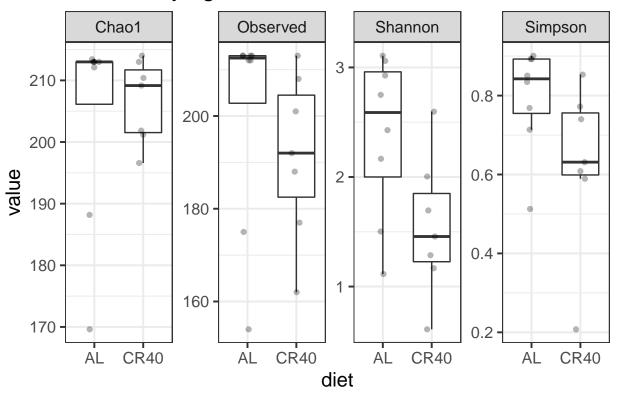
Alpha diversity is sensitive to sequencing depth, so I recommend performing rarefaction prior to running alpha diversity analysis so that the results aren't confounded by sequencing depth.

```
alpha.div.genus.rarefied.df <- physeq.bac.genus.filt %>%
    rarefy_even_depth(rngseed=101, replace=F, verbose=F) %>%
    estimate_richness(measures=c("Observed", "Chao1", "Shannon", "Simpson"))

alpha.div.genus.rarefied.df %>%
    merge(meta.df, by="row.names") %>%
    pivot_longer(c(Observed, Chao1, Shannon, Simpson), names_to="diversity") %>%

# CHANGEME to show desired comparisons
ggplot(aes(x=diet, y=value)) +
geom_boxplot(outlier.shape=NA) +
geom_jitter(width=0.1, alpha=0.3) +
facet_wrap(~diversity, ncol=4, scales="free_y") +
labs(title="After rarefying")
```

After rarefying

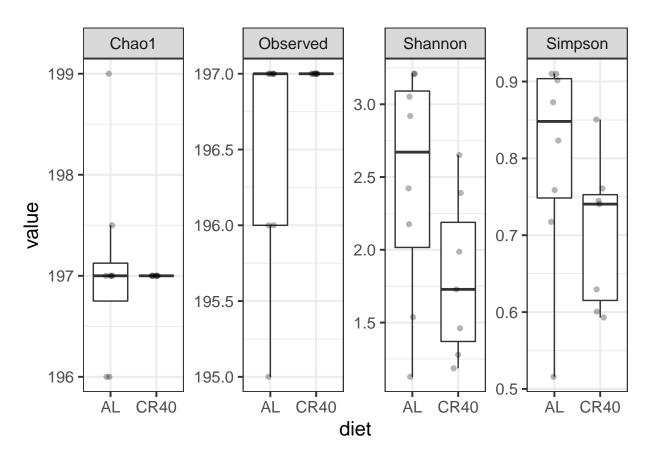


Species

```
alpha.div.species.df <- estimate_richness(
   physeq.bac.species.filt, measures=c("Observed", "Chao1", "Shannon", "Simpson"))

alpha.div.species.df %>%
   merge(meta.df, by="row.names") %>%
   pivot_longer(c(Observed, Chao1, Shannon, Simpson), names_to="diversity") %>%

# CHANGEME to show desired comparisons
ggplot(aes(x=diet, y=value)) +
geom_boxplot(outlier.shape=NA) +
geom_jitter(width=0.1, alpha=0.3) +
facet_wrap(~diversity, ncol=4, scales="free_y")
```



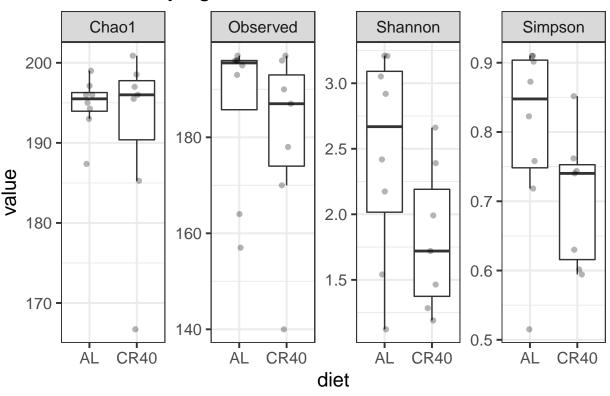
After rarefaction

```
alpha.div.species.rarefied.df <- physeq.bac.species.filt %>%
    rarefy_even_depth(rngseed=101, replace=F, verbose=F) %>%
    estimate_richness(measures=c("Observed", "Chao1", "Shannon", "Simpson"))

alpha.div.species.rarefied.df %>%
    merge(meta.df, by="row.names") %>%
    pivot_longer(c(Observed, Chao1, Shannon, Simpson), names_to="diversity") %>%

# CHANGEME to show desired comparisons
ggplot(aes(x=diet, y=value)) +
geom_boxplot(outlier.shape=NA) +
geom_jitter(width=0.1, alpha=0.3) +
facet_wrap(~diversity, ncol=4, scales="free_y") +
labs(title="After rarefying")
```

After rarefying



Barplots

Phylum

```
top.n.phyla <- data.frame(tax_table(physeq.bac.phylum))[
   names(sort(taxa_sums(physeq.bac.phylum), decreasing=T))[1:9], "phylum"]

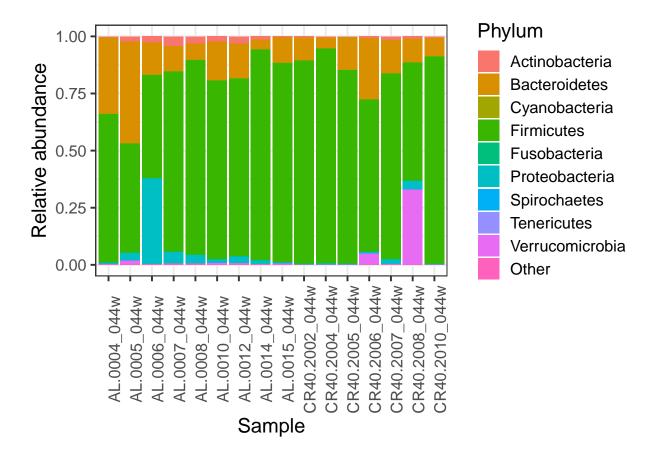
physeq.bac.phylum %>%

# compute relative abundance
transform_sample_counts(function(OTU) OTU/sum(OTU)) %>%

# convert to df
psmelt %>%

# aggregate less common taxa into "Other"
mutate(agg.phylum=fct_relevel(
   ifelse(phylum %in% top.n.phyla, phylum, "Other"), "Other", after=Inf)) %>%

# CHANGEME to show desired comparisons
ggplot(aes(x=Sample, y=Abundance, fill=agg.phylum)) +
geom_bar(stat="identity") +
labs(y="Relative abundance", fill="Phylum") +
theme(axis.text.x=element_text(angle=90))
```



```
top.n.genera <- data.frame(tax_table(physeq.bac.genus.filt))[
    names(sort(taxa_sums(physeq.bac.genus.filt), decreasing=T))[1:9], "genus"]

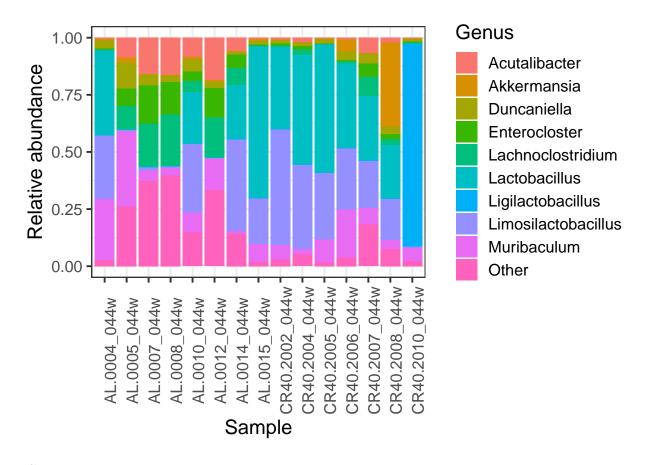
physeq.bac.genus.filt %>%

# compute relative abundance
    transform_sample_counts(function(OTU) OTU/sum(OTU)) %>%

# convert to df
    psmelt %>%

# aggregate less common taxa into "Other"
    mutate(agg.genus=fct_relevel(
        ifelse(genus %in% top.n.genera, genus, "Other"), "Other", after=Inf)) %>%

# CHANGEME to show desired comparisons
    ggplot(aes(x=Sample, y=Abundance, fill=agg.genus)) +
    geom_bar(stat="identity") +
    labs(y="Relative abundance", fill="Genus") +
    theme(axis.text.x=element_text(angle=90))
```



Species

Use genus_and_species annotation to improve the labels on the plot.

```
top.n.species <- data.frame(tax_table(physeq.bac.species.filt))[
   names(sort(taxa_sums(physeq.bac.species.filt), decreasing=T))[1:9], "genus_and_species"]

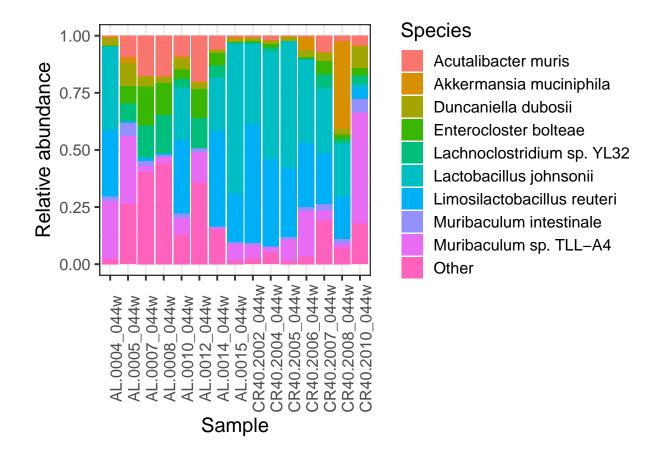
physeq.bac.species.filt %>%

# compute relative abundance
   transform_sample_counts(function(OTU) OTU/sum(OTU)) %>%

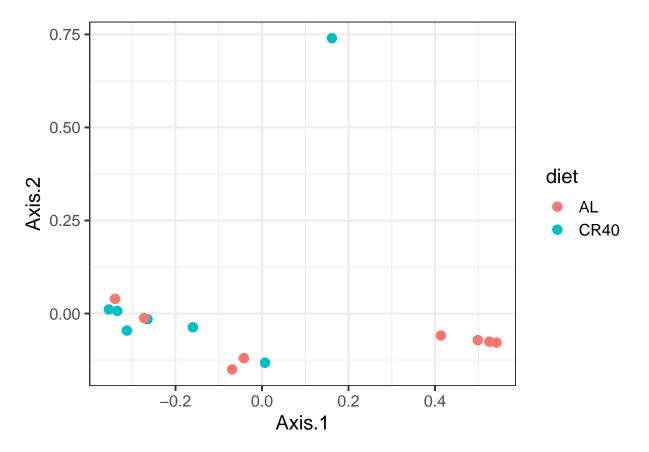
# convert to df
   psmelt %>%

# aggregate less common taxa into "Other"
   mutate(agg.species=fct_relevel(
    ifelse(genus_and_species %in% top.n.species, genus_and_species, "Other"), "Other", after=Inf)) %>%

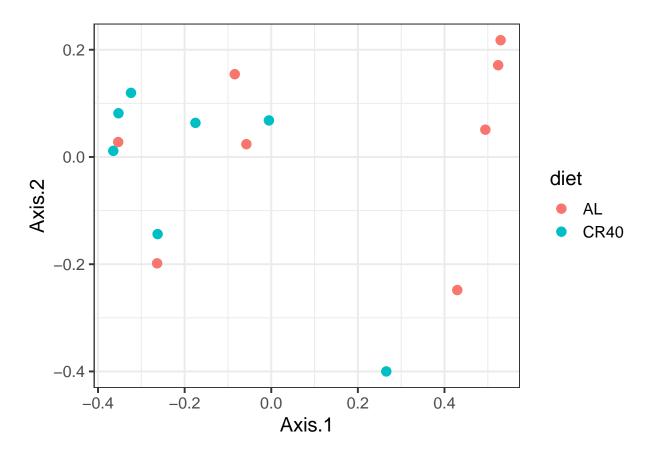
# CHANGEME to show desired comparisons
   ggplot(aes(x=Sample, y=Abundance, fill=agg.species)) +
   geom_bar(stat="identity") +
   labs(y="Relative abundance", fill="Species") +
   theme(axis.text.x=element_text(angle=90))
```



PCoA



Species



DESeq2 to find differential taxons

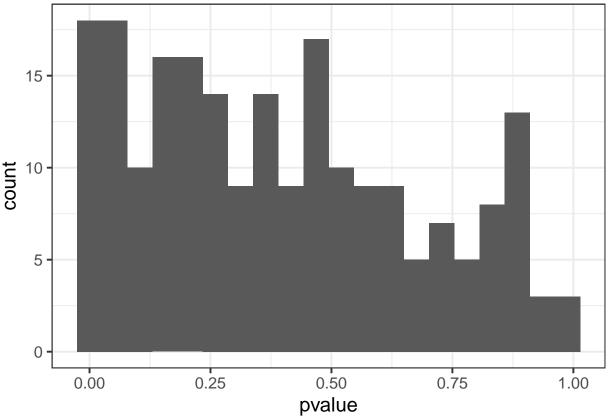
```
deseq.genus <- physeq.bac.genus.filt %>%
  # CHANGEME: update the design formula to perform desired comparisons
 phyloseq_to_deseq2(~diet)
## converting counts to integer mode
## Warning in DESeqDataSet(se, design = design, ignoreRank): some variables in
## design formula are characters, converting to factors
deseq.genus <- DESeq(deseq.genus)</pre>
## estimating size factors
## estimating dispersions
## gene-wise dispersion estimates
## mean-dispersion relationship
## -- note: fitType='parametric', but the dispersion trend was not well captured by the
      function: y = a/x + b, and a local regression fit was automatically substituted.
##
      specify fitType='local' or 'mean' to avoid this message next time.
## final dispersion estimates
## fitting model and testing
```

```
## -- replacing outliers and refitting for 4 genes
## -- DESeq argument 'minReplicatesForReplace' = 7
## -- original counts are preserved in counts(dds)

## estimating dispersions

## fitting model and testing

results(deseq.genus) %>% as.data.frame %>%
    ggplot(aes(pvalue)) +
    geom_histogram(bins=20)
```



• The shape of this histogram tells you whether to expect any significant results

```
results(deseq.genus) %>% as.data.frame %>%
  dplyr::filter(padj < 0.1)</pre>
```

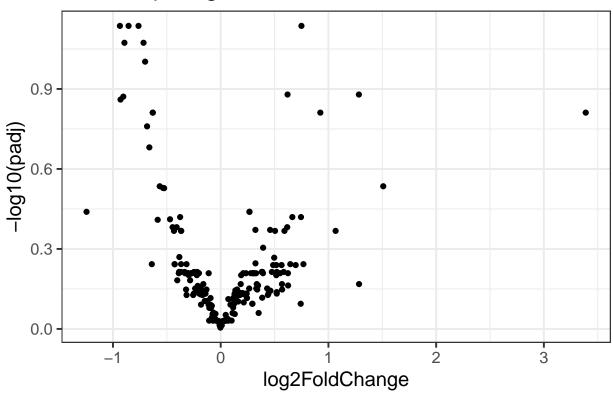
```
##
             baseMean log2FoldChange
                                          lfcSE
                                                     stat
                                                                pvalue
                                                                              padj
## 39488
            393.97495
                          -0.8936661 0.2941429 -3.038204 0.0023799284 0.08448746
## 2763056 1450.57113
                          -0.7016046 0.2385345 -2.941313 0.0032682398 0.09944787
## 35701
             17.61384
                          -0.7630718 0.2356969 -3.237514 0.0012057621 0.07301969
## 2583452
             46.21434
                          -0.7164464 0.2347089 -3.052489 0.0022695182 0.08448746
## 365349
             23.02017
                           0.7494517 0.2243861 3.340009 0.0008377557 0.07301969
## 128785
             15.30414
                          -0.8552258 0.2431065 -3.517907 0.0004349655 0.07301969
## 81412
             13.61889
                          -0.9356860 0.2923441 -3.200633 0.0013712618 0.07301969
```

- 7 results with adjusted p-value < 0.1
- The adjusted p-value threshold is somewhat arbitrary

```
results(deseq.genus) %>% as.data.frame %>%
ggplot(aes(x=log2FoldChange, y=-log10(padj))) +
```

```
geom_point() +
labs(title="Volcano plot, genera")
```

Volcano plot, genera



Species

```
deseq.species <- physeq.bac.species.filt %>%

# CHANGEME: update the design formula to perform desired comparisons
phyloseq_to_deseq2(~diet)

## converting counts to integer mode

## Warning in DESeqDataSet(se, design = design, ignoreRank): some variables in

## design formula are characters, converting to factors

deseq.species <- DESeq(deseq.species)

## estimating size factors

## estimating dispersions

## gene-wise dispersion estimates

## mean-dispersion relationship

## -- note: fitType='parametric', but the dispersion trend was not well captured by the

## function: y = a/x + b, and a local regression fit was automatically substituted.

## specify fitType='local' or 'mean' to avoid this message next time.

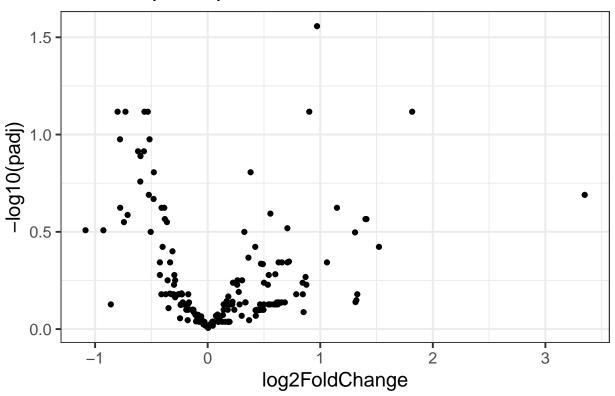
## final dispersion estimates</pre>
```

```
## fitting model and testing
## -- replacing outliers and refitting for 2 genes
## -- DESeq argument 'minReplicatesForReplace' = 7
## -- original counts are preserved in counts(dds)
## estimating dispersions
## fitting model and testing
results(deseq.species) %>% as.data.frame %>%
  ggplot(aes(pvalue)) +
  geom_histogram(bins=20)
   15
   10
    5
    0
                            0.25
                                              0.50
                                                                0.75
                                                                                 1.00
          0.00
                                           pvalue
results(deseq.species) %>% as.data.frame %>%
  dplyr::filter(padj < 0.1)</pre>
            baseMean log2FoldChange
##
                                        lfcSE
                                                    stat
                                                               pvalue
## 1912897 679.75388
                          0.9713758 0.2551635 3.806877 0.0001407329 0.02772438
                         -0.7303005 0.2433224 -3.001370 0.0026876759 0.07631305
## 39488
           375.83986
## 351091
           386.04652
                         -0.5315480 0.1770845 -3.001662 0.0026850977 0.07631305
## 1584
            36.09639
                          0.9027695 0.2943759 3.066723 0.0021641909 0.07631305
## 414771
            93.79130
                          1.8166514 0.5773366 3.146607 0.0016517677 0.07631305
                         -0.5618419 0.1873638 -2.998667 0.0027116313 0.07631305
## 1685
            72.10095
## 128785
            13.34972
                         -0.8000896 0.2445831 -3.271239 0.0010707740 0.07631305
results(deseq.species) %>% as.data.frame %>%
  ggplot(aes(x=log2FoldChange, y=-log10(padj))) +
```

geom_point() +

labs(title="Volcano plot, species")

Volcano plot, species



Save for reproducibility

```
Sys.Date()
## [1] "2022-12-15"
sessionInfo()
## R version 4.2.0 (2022-04-22)
## Platform: x86_64-apple-darwin17.0 (64-bit)
## Running under: macOS Big Sur/Monterey 10.16
##
## Matrix products: default
## BLAS:
          /Library/Frameworks/R.framework/Versions/4.2/Resources/lib/libRblas.0.dylib
## LAPACK: /Library/Frameworks/R.framework/Versions/4.2/Resources/lib/libRlapack.dylib
##
## locale:
## [1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
## attached base packages:
## [1] stats4
                           graphics grDevices utils
                 stats
                                                         datasets methods
## [8] base
##
## other attached packages:
   [1] speedyseq_0.5.3.9018
                                    DESeq2_1.36.0
   [3] SummarizedExperiment_1.26.1 Biobase_2.56.0
##
   [5] MatrixGenerics_1.8.1
                                    matrixStats_0.62.0
```

```
[7] GenomicRanges 1.48.0
                                     GenomeInfoDb 1.32.4
##
  [9] IRanges_2.30.1
                                     S4Vectors_0.34.0
## [11] BiocGenerics 0.42.0
                                     phyloseq 1.40.0
                                     stringr_1.4.1
## [13] forcats_0.5.2
## [15] dplyr_1.0.10
                                     purrr_0.3.4
## [17] readr 2.1.2
                                     tidyr 1.2.1
## [19] tibble 3.1.8
                                     ggplot2_3.3.6
## [21] tidyverse_1.3.2
##
## loaded via a namespace (and not attached):
     [1] googledrive_2.0.0
                                 colorspace_2.0-3
                                                        ellipsis_0.3.2
     [4] XVector_0.36.0
##
                                 fs 1.5.2
                                                        rstudioapi_0.14
##
     [7] farver_2.1.1
                                 bit64_4.0.5
                                                        AnnotationDbi_1.58.0
   [10] fansi_1.0.3
                                 lubridate_1.8.0
##
                                                        xm12_1.3.3
    [13] codetools_0.2-18
                                 splines_4.2.0
                                                        cachem_1.0.6
##
##
    [16] geneplotter_1.74.0
                                 knitr_1.40
                                                        ade4_1.7-19
##
    [19] jsonlite_1.8.3
                                 broom_1.0.1
                                                        annotate_1.74.0
##
   [22] cluster 2.1.4
                                 dbplyr_2.2.1
                                                        png 0.1-7
   [25] compiler_4.2.0
                                 httr_1.4.4
##
                                                        backports_1.4.1
    [28] assertthat 0.2.1
                                 Matrix 1.5-1
                                                        fastmap 1.1.0
##
    [31] gargle_1.2.1
                                 cli_3.4.1
                                                        htmltools_0.5.3
   [34] tools 4.2.0
                                 igraph_1.3.4
                                                        gtable_0.3.1
##
    [37] glue_1.6.2
                                 GenomeInfoDbData_1.2.8 reshape2_1.4.4
##
    [40] Rcpp 1.0.9
                                                        vctrs 0.5.0
##
                                 cellranger 1.1.0
##
   [43] Biostrings_2.64.1
                                 rhdf5filters 1.8.0
                                                        multtest 2.52.0
   [46] ape_5.6-2
                                 nlme_3.1-159
                                                        iterators 1.0.14
##
    [49] xfun_0.34
                                 rvest_1.0.3
                                                        lifecycle_1.0.3
    [52] XML_3.99-0.10
##
                                 googlesheets4_1.0.1
                                                        zlibbioc_1.42.0
##
   [55] MASS_7.3-58.1
                                 scales_1.2.1
                                                        hms_1.1.2
    [58] parallel_4.2.0
                                 biomformat_1.24.0
                                                        rhdf5_2.40.0
##
    [61] RColorBrewer_1.1-3
                                 yaml_2.3.6
                                                        memoise_2.0.1
##
    [64] RSQLite_2.2.17
                                 stringi_1.7.8
                                                        highr_0.9
   [67] genefilter_1.78.0
                                 foreach_1.5.2
                                                        permute_0.9-7
   [70] BiocParallel_1.30.3
                                                        pkgconfig_2.0.3
                                 rlang_1.0.6
##
    [73] bitops 1.0-7
                                 evaluate 0.17
                                                        lattice 0.20-45
   [76] Rhdf5lib_1.18.2
##
                                                        bit 4.0.4
                                 labeling_0.4.2
  [79] tidyselect 1.1.2
                                 plyr_1.8.7
                                                        magrittr 2.0.3
##
   [82] R6_2.5.1
                                 generics_0.1.3
                                                        DelayedArray_0.22.0
    [85] DBI_1.1.3
                                 pillar_1.8.1
                                                        haven_2.5.1
##
  [88] withr_2.5.0
                                                        KEGGREST_1.36.3
##
                                 mgcv_1.8-40
                                                        modelr 0.1.9
  [91] survival 3.4-0
                                 RCurl 1.98-1.8
   [94] crayon_1.5.1
                                 utf8 1.2.2
                                                        tzdb 0.3.0
##
  [97] rmarkdown 2.17
                                 locfit 1.5-9.6
                                                        grid 4.2.0
## [100] readxl_1.4.1
                                 data.table_1.14.4
                                                        blob_1.2.3
## [103] vegan_2.6-2
                                 reprex_2.0.2
                                                        digest_0.6.30
## [106] xtable_1.8-4
                                 munsell_0.5.0
```